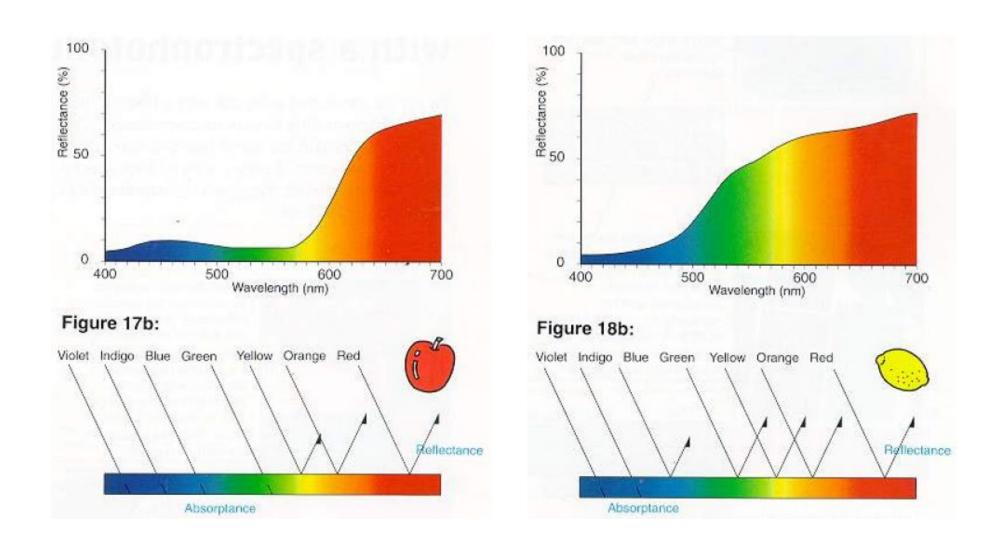
CMPG-767 Digital Image Processing

PROCESSING OF COLOR IMAGES

- Color can only exist when three components are present: a viewer, an object, and light
- When white light hits an object, it selectively blocks some colors and reflects others
- Only the reflected colors contribute to the viewer's perception

of color



- Primary colors of light: Red, Green, Blue
 - The specific wavelength values to the three primary colors
 - ➤Blue= 435.8 nm, green= 546.1 nm, red= 700 nm
- Secondary colors of light: Cyan, Magenta, Yellow
- The primary colors should be added to produce secondary colors of Light
 - Magenta (Red+Blue)
 - Cyan (Green+ Blue)
 - Yellow (Red+ Green)

- Primary colors of pigments or colorants
 - cyan, magenta, yellow
 - A primary color of pigments is defined as one that <u>subtracts or</u> <u>absorbs</u> a primary color of light and reflects or transmits the other two
- Secondary colors of pigments or colorants
 - red, green, blue
 - Combination of the three pigment primaries, or a secondary with its opposite primary, produces black

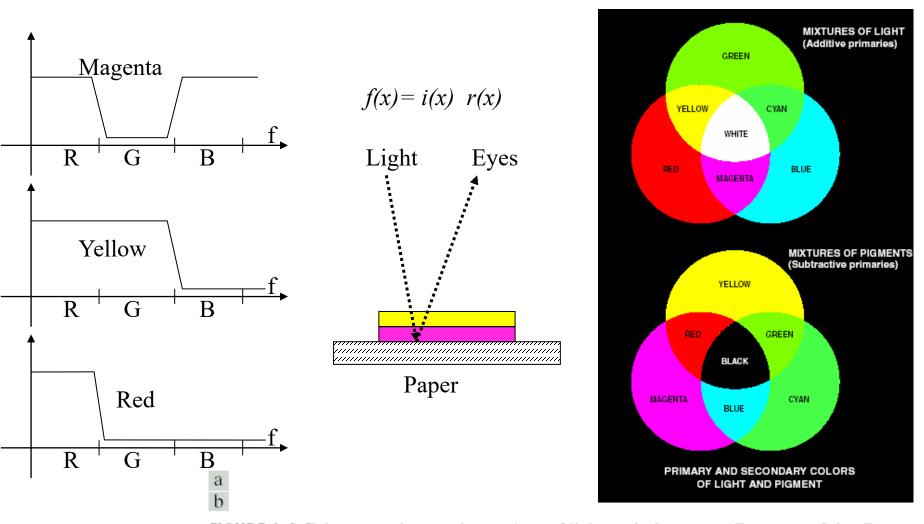
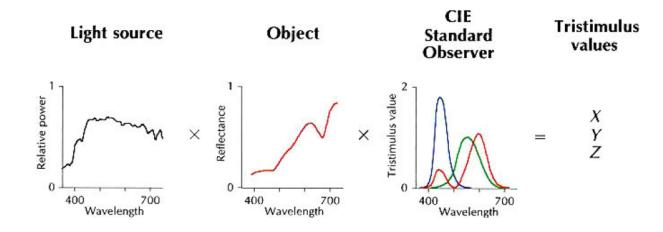


FIGURE 6.4 Primary and secondary colors of light and pigments. (Courtesy of the General Electric Co., Lamp Business Division.)

- Characteristics of colors
 - Brightness
 - The chromatic notion of intensity
 - Hue
 - An attribute associated with the dominant wavelength in a mixture of light waves
 - Representing dominant color as perceived by an observer
 - Saturation
 - Referring to relative purity or the amount of white mixed with a hue
 - Saturation is inversely proportional to the amount of white light

- Hue and saturation taken together are called chromaticity
 - A color may be characterized by its brightness and chromaticity
 - The amounts of red, green, and blue needed to form any particular color are called the tristimulus values

• Denoted X(red), Y(green), and Z(blue)
$$x = \frac{X}{X + Y + Z}, \quad y = \frac{Y}{X + Y + Z}, \quad z = \frac{Z}{X + Y + Z}$$
$$x + y + z = 1 \ , \quad X = \frac{x}{y}Y, Z = \frac{z}{y}Y$$



Human Perception

Detailed experimental evidences has established that the 6 to 7 million cones in the human eye can be divided into three principal sensing categories, corresponding roughly to red, green and blue

Approximately

65% of all cones are sensitive to Red Light,

33% are sensitive to Green Light

2% are sensitive to Blue Light

Absorption of Light by red, green and blue cones in Human Eye

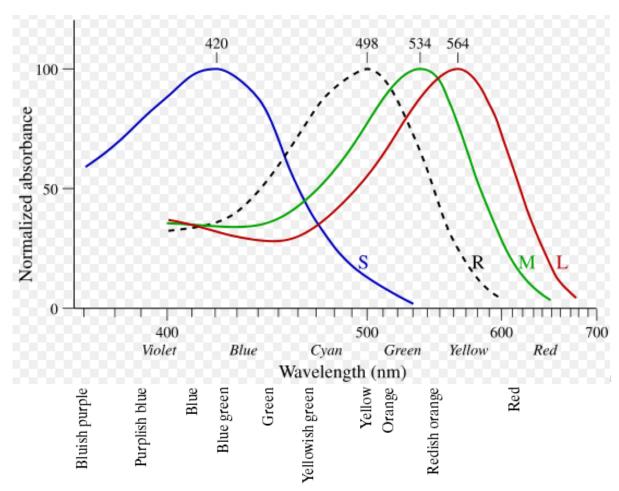


FIGURE 6.3 Absorption of light by the red, green, and blue cones in the human eye as a function of wavelength.

Mixing the three primaries or a secondary with its opposite primary colors in the right intensities produces white light

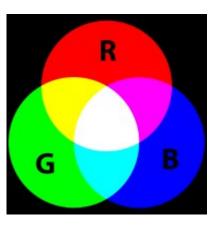


Color Models

- The purpose of a color model (also called Color Space or Color System) is to facilitate the specification of colors in some standard way
- A color model is a specification of a coordinate system and a subspace within that system where each color is represented by a single point
- Color Models

RGB (Red, Green, Blue)
CMY (Cyan, Magenta, Yellow)
HSI (Hue, Saturation, Intensity)
YIQ (Luminance,In phase, Quadrature)
YUV (Y' stands for the luma component (the brightness) and U and V are the chrominance (color) components)

- RGB (Red Green Blue) uses additive color mixing, because it describes what kind of light needs to be emitted to produce a given color.
- Additive color mixing: Three overlapping light bulbs in a vacuum, adding together to create white



- In the RGB space, its components measure the intensity and chrominance of light
- The actual information stored in the digital image data is the intensity information in each channel
- If a digital color image is represented using 8 bit/component precision, then the 24-bit RGB model may represent
 - $256 \times 256 \times 256 \approx 16.7$ million colors

- Each color is represented in its primary color components Red, Green and Blue
- This model is based or Cartesian Coordinate System
- This is the model used for active displays such as television and computer screens

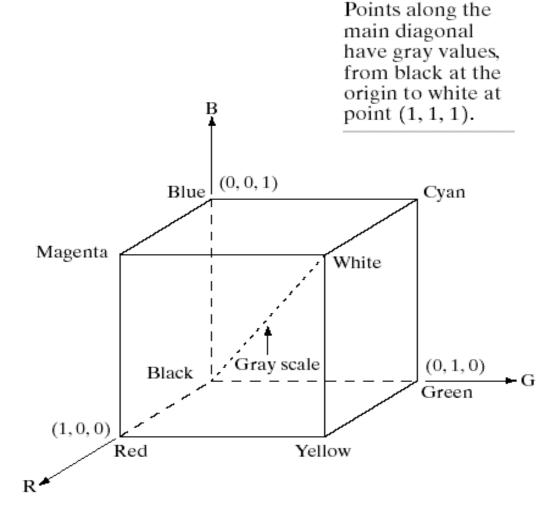
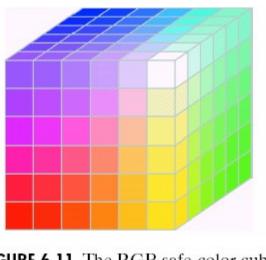


FIGURE 6.7

Schematic of the RGB color cube.

- The RGB model is usually represented by a unit cube with one corner located at the origin of a three-dimensional color coordinate system, the axes being labeled R, G, B, and having a range of values [0, 1].
- The origin (0, 0, 0) is considered black and the diagonally opposite corner (1, 1, 1) is called white. The line joining black to white represents a gray scale and has equal components of R, G, B.





It is important to note that not all possible 8-bit gray colors are included in the **216 safe** colors

- The set of *safe RGB colors*
 - A subset of colors that are likely to be reproduced faithfully, reasonably independently of viewer hardware capabilities
 - They are also called the set of all-system-safe colors, safe Web colors, safe browser colors
 - 216 safe colors in RGB color model
 - On the assumption that 256 colors is the minimum number of colors that can be reproduced faithfully by any system, forty of these 256 colors are known to be processed differently by various operating systems, leaving only 216 colors that are common to most systems

Generating RGB image

Number System	Color Equivalents					
Hex	00	33	66	99	CC	FF
Decimal		51	102	153	204	255

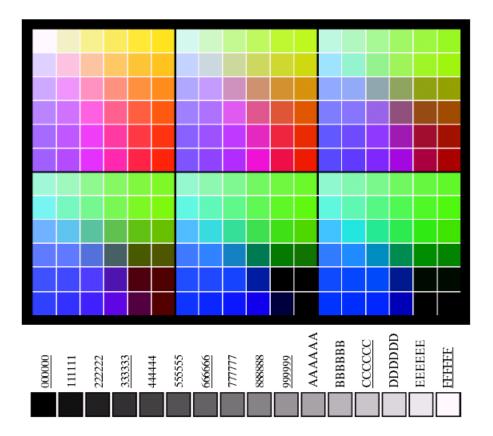


TABLE 6.1Valid values of each RGB component in a safe color.

a

FIGURE 6.10

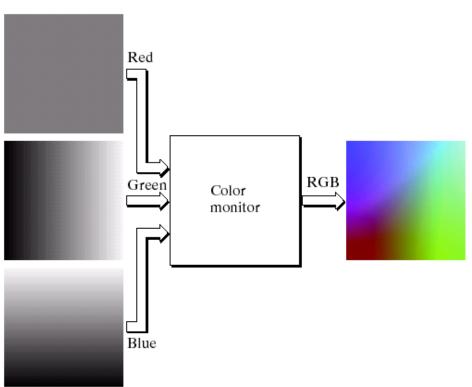
(a) The 216 safe RGB colors.
(b) All the grays in the 256-color RGB system (grays that are part of the safe color group are shown underlined).

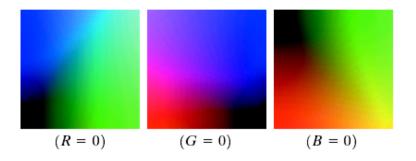
Generating the RGB image

> a b

FIGURE 6.9

(a) Generating the RGB image the cross-sectic color plane (127, G, B). (b) The three hidden surface planes in the cocube of Fig. 6.8.





RGB Space and YUV Space

 RGB pixel intensities can be transformed to the YUV space as follows

$$\begin{pmatrix} Y \\ U \\ V \end{pmatrix} = \begin{pmatrix} 0.299 & 0.587 & 0.114 \\ -0.14713 & -0.28886 & 0.436 \\ 0.615 & -0.51499 & -0.10001 \end{pmatrix} \begin{pmatrix} R \\ G \\ B \end{pmatrix}$$

Where Y is a luminosity (luma) information (channel), U and V are chromatic (differences U=B-R-G, V=R-G-B) channels

RGB Space and YUV Space

 RGB pixel intensities can be restored from the YUV space as follows

$$\begin{pmatrix} R \\ G \\ B \end{pmatrix} = \begin{pmatrix} 1 & 0 & 1.13983 \\ 1 & -0.39465 & -0.5806 \\ 1 & 2.03211 & 0 \end{pmatrix} \begin{pmatrix} Y \\ U \\ V \end{pmatrix}$$

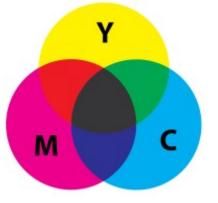
 Where Y is a luminosity information (channel), U and V are chromatic (differences U=B-R-G, V=R-G-B) channels

YUV space and JPEG compression

- YUV space is used in JPEG image compression
- To apply JPEG compression, an RGB image shall be transformed into an YUV image
- Then, JPEG compression is applied to the YUV channels. This
 makes it possible to reach a higher compression rate, because
 the U and V channels can easily be compressed with a
 significantly higher rate (up to 90% without losing any
 significant information) rather than any of RGB channels

CMYK color model

- CMYK (Cyan Magenta Yellow Key) uses subtractive color mixing used in the printing process, because it describes what kind of inks need to be applied so the light reflected from the substrate and through the inks produces a given color.
- Subtractive color mixing: Three splotches of paint on white paper, subtracting together to turn the paper black



CMY and CMYK Color Model

- Cyan, magenta, and yellow are the secondary colors with respect to the primary colors of red, green, and blue.
- In this model, colors are formed by subtraction, where adding different pigments causes various colors not to be reflected and thus not to be seen. Here, white is the absence of colors, and black is the sum of all of them. This is generally the model used for printing.
- Most devices that deposit color pigments on paper (such as Color Printers and Copiers)
 requires CMY data input or perform RGB to CMY conversion internally

$$\begin{bmatrix} C \\ M \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} - \begin{bmatrix} R \\ G \end{bmatrix}$$

$$\begin{bmatrix} Y \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} - \begin{bmatrix} B \end{bmatrix}$$

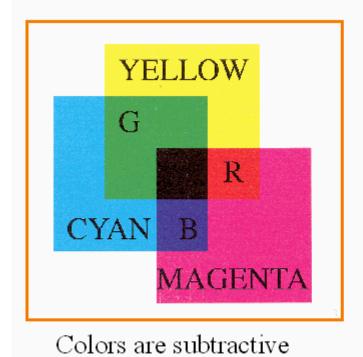
CMY and CMYK Color Model

CMY is a Subtractive Color Model

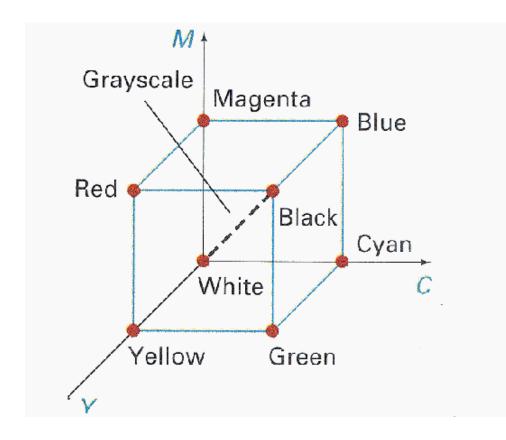
$$\begin{bmatrix} C \\ M \\ Y \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ - \begin{bmatrix} G \\ B \end{bmatrix}$$

- Equal amounts of Pigment primaries (Cyan, Magenta and Yellow) should produce Black
- In practice combining these colors for printing produces a "Muddy-Black" color
- So in order to produce "True-Black" a fourth color "Black" is added giving rise to CMYK model

CMYK color model

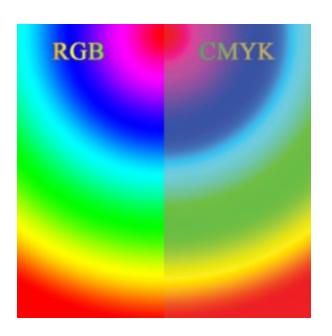


C	M	Y	Color
0.0	0.0	0.0	White
1.0	0.0	0.0	Cyan
0.0	1.0	0.0	Magenta
0.0	0.0	1.0	Yellow
1.0	1.0	0.0	Blue
1.0	0.0	1.0	Green
0.0	1.0	1.0	Red
1.0	1.0	1.0	Black
0.5	0.0	0.0	
1.0	0.5	0.5	
1.0	0.5	0.0	



RGB vs. CMYK

A comparison of RGB and CMYK color models. This image demonstrates the
difference between how colors will look on a computer monitor (RGB) compared to
how they will reproduce in a CMYK print process



RGB vs. CMYK

 CMYK color space is used primarily for color printing (the K component characterizes specific properties of a particular printer)

 RGB color space is used for storing color images, their processing (filtering, enhancement), and displaying

> Hue (dominant colour seen)

- Wavelength of the pure colour observed in the signal.
- Distinguishes red, yellow, green, etc.
- More the 400 hues can be seen by the human eye.

➤ Saturation (degree of dilution)

- Inverse of the quantity of "white" present in the signal. A pure colour has 100% saturation, the white and grey have 0% saturation.
- Distinguishes red from pink, marine blue from royal blue, etc.
- About 20 saturation levels are visible per hue.

> Intensity

Distinguishes the gray levels.

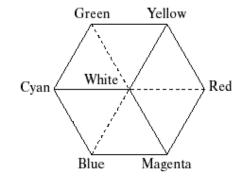
- Usefulness
 - The intensity is decoupled from the color information

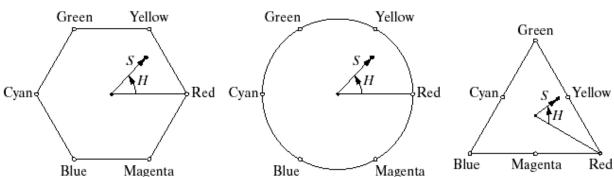
 The hue and saturation are intimately related to the way in which human beings perceive color

 An ideal tool for developing image procession algorithms based on some of the color sensing properties of the human visual system

- Hue and saturation in the HSI color model
 - Hue Angle from the red axis

Saturation - Length of the vector





a b c d

FIGURE 6.13 Hue and saturation in the HSI color model. The dot is an arbitrary color point. The angle from the red axis gives the hue, and the length of the vector is the saturation. The intensity of all colors in any of these planes is given by the position of the plane on the vertical intensity axis.

- Converting colors from RGB to HIS
 - Hue component

$$H = \begin{cases} \theta & \text{if } B \le G \\ 360 - \theta & \text{if } B > G \end{cases}$$

$$H = \begin{cases} \theta \text{ if } B \le G \\ 360 - \theta \text{ if } B > G \end{cases} \qquad \theta = \cos^{-1} \left[\frac{\frac{1}{2} \{ (R - G) + (R - B) \}}{\{ (R - G)^2 + (R - B)(G - B) \}^{1/2}} \right]$$

Saturation component

$$S = 1 - \frac{3}{(R+G+B)}[\min(R,G,B)]$$

Intensity component

$$I = \frac{1}{3}(R + G + B)$$

RGB values have been normalized to the range [0,1]

Angle θ is measured with respect to the red axis

Hue can be normalized to the range [0, 1] by dividing by 360°c

Converting colors from HSI to RGB

- •Three sectors of interest, corresponding to the 1200 intervals in the separation of primaries
 - **GB** sector $(120^{\circ} \le H < 240^{\circ})$ o BR sector $(240^{\circ} \le H < 360^{\circ})$

• BR sector
$$(240^{\circ} \le H < 360^{\circ})$$

$$H = H - 120^{\circ}$$

$$R = I(1-S)$$

$$G = I\left[1 + \frac{S\cos H}{\cos(60^{\circ} - H)}\right]$$

$$B = 3I - (R+G)$$

$$H = H - 240^{\circ}$$

$$G = I(1-S)$$

$$B = I\left[1 + \frac{S\cos H}{\cos(60^{\circ} - H)}\right]$$

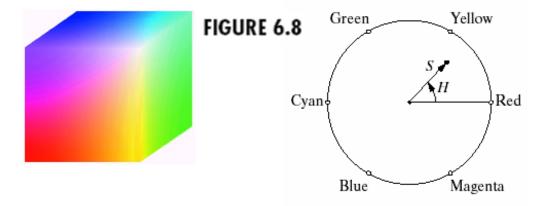
$$R = 3I - (G+B)$$

Converting colors from HSI to RGB

Example 6.2

a b c

The HSI values corresponding to the image of the RGB color cube



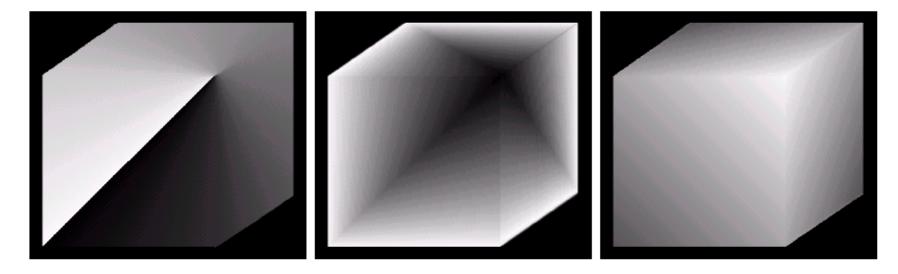
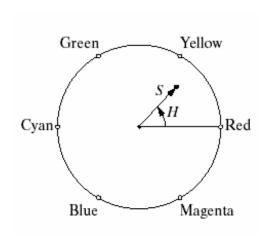


FIGURE 6.15 HSI components of the image in Fig. 6.8. (a) Hue, (b) saturation, and (c) intensity images.

Converting colors from HSI to RGB

Manipulating HSI component images



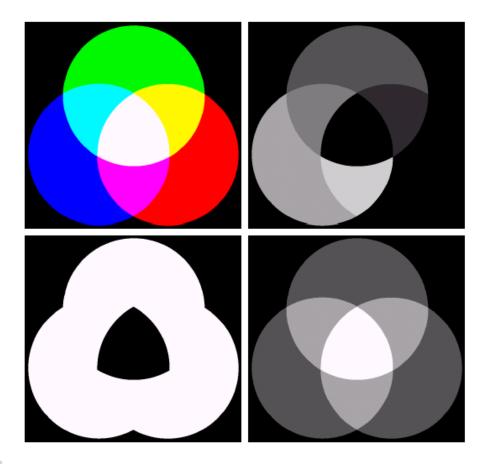
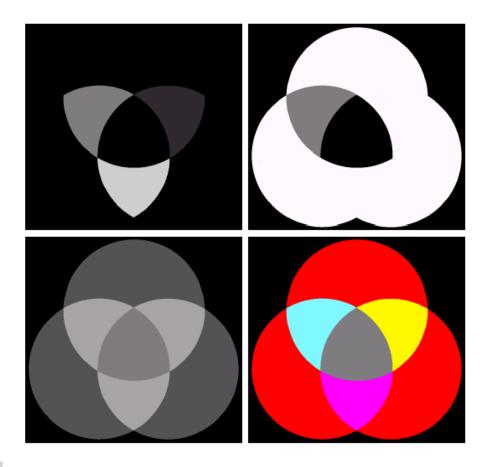




FIGURE 6.16 (a) RGB image and the components of its corresponding HSI image: (b) hue, (c) saturation, and (d) intensity.

Converting colors from HSI to RGB



a b c d

FIGURE 6.17 (a)–(c) Modified HSI component images. (d) Resulting RGB image. (See Fig. 6.16 for the original HSI images.)

Processing of color images

- Any kind of color image processing can be implemented in two alternative ways:
 - Each color channel should be processed separately, in general using even different filters with different parameters. Moreover, one or two color channels may remain unprocessed
 - A luminosity information can be extracted from an RGB image; then, after its processing, it can be inserted back into the image

Processing of color images

- A pixel at (x,y) is a vector in the color space
 - RGB color space

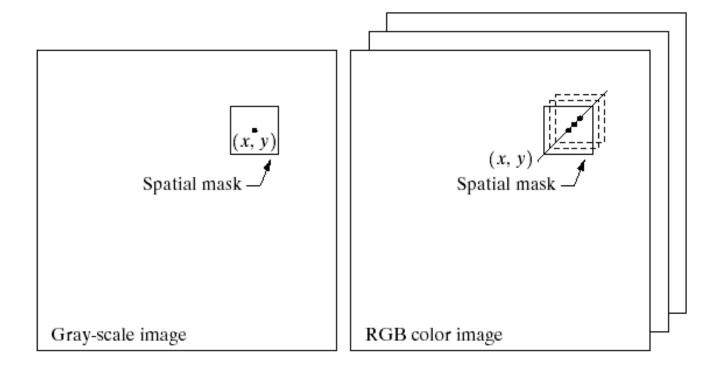
$$\mathbf{c}(x,y) = \begin{bmatrix} R(x,y) \\ G(x,y) \\ B(x,y) \end{bmatrix}$$

Processing of color images in spatial domain

a b

FIGURE 6.29

Spatial masks for gray-scale and RGB color images.



• Similar to gray scale transformation

$$-g(x,y)=T[f(x,y)]$$

Color transformation

$$S_i = T_i(r_1, r_2, ..., r_n), i = 1, 2, ..., n$$

$$g(x,y) \qquad \qquad f(x,y)$$

$$S_1 \leftarrow T_1 \qquad f_1$$

$$S_2 \leftarrow T_2 \qquad \dots$$

$$S_n \qquad T_n \qquad f_n$$

Formulation

$$g(x,y) = T[f(x,y)]$$

- f(x, y) is a color input image
- g(x, y) is the transformed or processed
- T is an operator
- Transformation or color mapping functions

$$S_i = T_i(r_1, r_2, ..., r_n)$$

- n is the number of color components
- The color components of f(x, y) and g(x, y) at any point
- $\{T_1, T_2, ..., T_n\}$ is a set of transformation

Some operations are better suited to specific models.

$$g(x,y) = kf(x,y)$$

• In the HIS color space, this can be done with the simple transformation

$$s_3 = kr_3$$
, where $0 < k < 1$

• In the RGB color space, three components must be transformed:

$$s_i = kr_i$$
 $i = 1,2,3$.

The CMY space requires a similar set of tranfomations:

$$S_i = kr_i + (1-k), i = 1,2,3.$$

Full color

Hue

Color Transformations

Black

FIGURE 6.30 A full-color image and its various color-space components. (Original image courtesy of Med-Data Interactive.)



Saturation

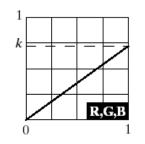
Intensity

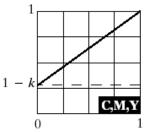
a b

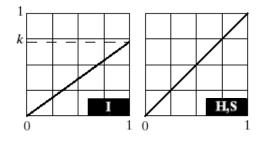
FIGURE 6.31 Adjusting the intensity of an image using color transformations. (a) Original image. (b) Result of decreasing its intensity by 30% (i.e., letting k = 0.7). (c)–(e) The required RGB, CMY, and HSI transformation functions. (Original image courtesy of MedData Interactive.)











• It is important that to note that each transformation defined in Eqs.(6.5-4) through (6.5-6) depends only on one component within its color space

Histogram Processing

- Histogram equalization
 - A transformation that seeks to produce an image with a uniform histogram of intensity values
- Color images
 - Composed of multiple components
 - The gray-scale technique to more than one component and/or histogram
 - A more logical approach
 - To spread the color intensities uniformly, leaving the colors themselves(e.g., hues) unchanged.
 - HSI color space is ideally suited to this type approach.

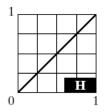
Histogram Processing

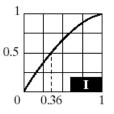
a b c d

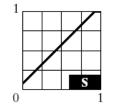
FIGURE 6.37

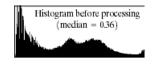
Histogram equalization (followed by saturation adjustment) in the HSI color space.

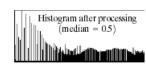
















Low-pass fileting (Smoothing) and Sharpening

 The next step beyond transforming each pixel of a color image without regard to its neighbors (as in the previous section) is to modify its value based on the characteristics of the surrounding pixels

 The basics of neighborhood processing are illustrated within the context of color image smoothing and sharpening.

- The basics of neighborhood processing
 - As a spatial filtering operation in which the coefficients of the filtering mask are all 1's
 - The mask is slid across the image to be smoothed
 - Each pixel is replaced by the average of the pixels in the neighborhood defined by the mask

$$\overline{c}(x,y) = \begin{bmatrix}
\frac{1}{K} \sum_{(x,y) \in S_{xy}} R(x,y) \\
\frac{1}{K} \sum_{(x,y) \in S_{xy}} G(x,y) \\
\frac{1}{K} \sum_{(x,y) \in S_{xy}} B(x,y)
\end{bmatrix}$$

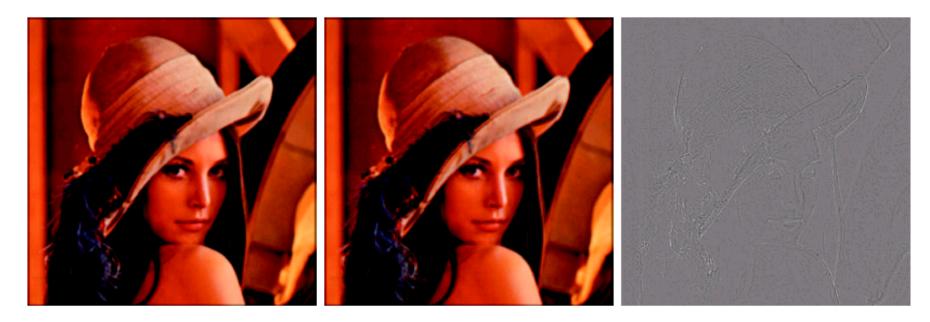


FIGURE 6.38

- (a) RGB image. (b) Red component image. (c) Green
- component. (d) Blue
- component.



FIGURE 6.39 HSI components of the RGB color image in Fig. 6.38(a). (a) Hue. (b) Saturation. (c) Intensity.



a b c

FIGURE 6.40 Image smoothing with a 5×5 averaging mask. (a) Result of processing each RGB component image. (b) Result of processing the intensity component of the HSI image and converting to RGB. (c) Difference between the two results.

Color Image Sharpening

Use the Laplacian (same as smoothing)



a b c

FIGURE 6.41 Image sharpening with the Laplacian. (a) Result of processing each RGB channel. (b) Result of processing the intensity component and converting to RGB. (c) Difference between the two results.

Processing of color images

 Separate processing of color channels is reasonable if, for example these channels are corrupted by different noises or if it is necessary to correct or enhance colors

 An intensity channel in the HIS space or luminosity channel in YUV space can be processed if only the luminosity information is corrupted

- By gradient operators
- The gradient discussed in section 3.7.3 is not defined for vector for quantities

$$\nabla f = mag(\nabla \mathbf{f}) = [G_{x}^{2} + G_{y}^{2}]^{1/2} = [(\frac{\partial f}{\partial x})^{2} + (\frac{\partial f}{\partial y})^{2}]^{1/2}$$

 Computing the gradient on individual images and then using the results to form a color image will lead to erroneous results

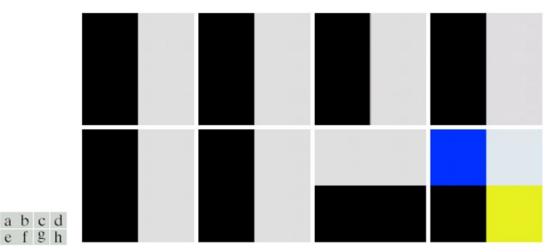


FIGURE 6.45 (a)–(c) R, G, and B component images and (d) resulting RGB color image. (f)–(g) R, G, and B component images and (h) resulting RGB color image.

- Definition of the gradient applicable to vector quantities: method by Di Zenzo
 - The gradient(magnitude and direction)of the vector c

$$\mathbf{u} = \frac{\partial R}{\partial x} \mathbf{r} + \frac{\partial G}{\partial x} \mathbf{g} + \frac{\partial B}{\partial x} \mathbf{b}$$

$$\mathbf{v} = \frac{\partial R}{\partial y} \mathbf{r} + \frac{\partial G}{\partial y} \mathbf{g} + \frac{\partial B}{\partial y} \mathbf{b}$$

$$g_{xx} = \mathbf{u} \cdot \mathbf{u} = \mathbf{u}^T \mathbf{u} = \left| \frac{\partial R}{\partial x} \right|^2 + \left| \frac{\partial G}{\partial x} \right|^2 + \left| \frac{\partial B}{\partial x} \right|^2$$

$$g_{yy} = \mathbf{v} \cdot \mathbf{v} = \mathbf{v}^T \mathbf{v} = \left| \frac{\partial R}{\partial y} \right|^2 + \left| \frac{\partial G}{\partial y} \right|^2 + \left| \frac{\partial B}{\partial y} \right|^2$$

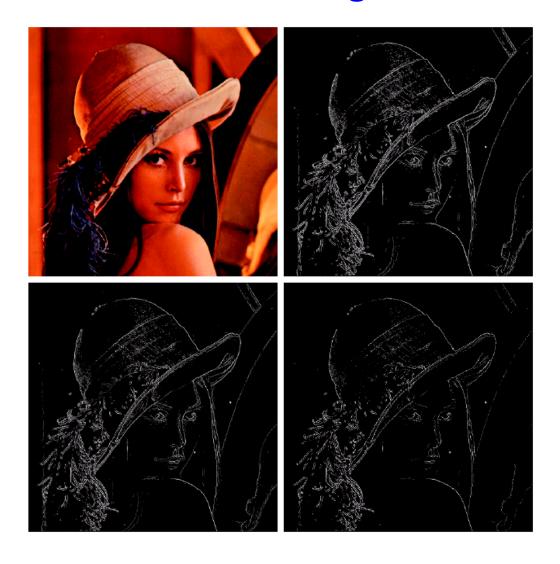
$$g_{xy} = \mathbf{u} \cdot \mathbf{v} = \mathbf{u}^T \mathbf{v} = \frac{\partial R}{\partial x} \frac{\partial R}{\partial y} + \frac{\partial G}{\partial x} \frac{\partial G}{\partial y} + \frac{\partial B}{\partial x} \frac{\partial B}{\partial y}$$

The direction of maximum rate of change of

$$\theta = \frac{1}{2} \tan^{-1} \left[\frac{2g_{xy}}{(g_{xx} - g_{yy})} \right]$$

• The value of the rate of change at (x, y)

$$F(\theta) = \left\{ \frac{1}{2} [(g_{xx} + g_{yy}) + (g_{xx} - g_{yy}) \cos 2\theta + 2g_{xy} \sin 2\theta] \right\}^{\frac{1}{2}}$$



a t

FIGURE 6.46

- (a) RGB image.
- (b) Gradient computed in RGB color vector space.
- (c) Gradients computed on a per-image basis and then added. (d) Difference between (b)

and (c).



FIGURE 6.47 Component gradient images of the color image in Fig. 6.46. (a) Red component, (b) green component, and (c) blue component. These three images were added and scaled to produce the image in Fig. 6.46(c).

a b c

Noise in Color Image

- The noise content of a color image
 - The same characteristics in each color channel
 - Possible for color channels to be affected differently by noise
- Different noise level
 - Caused by differences in the relative strength of illumination available to each of the color channels
 - For example(use of a red filter : CCD camera)
 - CCD sensors are noisier at lower levels of illumination

Noise in Color Image



a b c d

FIGURE 6.48
(a)–(c) Red,
green, and blue
component
images corrupted
by additive
Gaussian noise of
mean 0 and
variance 800.
(d) Resulting
RGB image.
[Compare (d)
with Fig. 6.46(a).]

Noise in Color Image

- The hue and saturation components are degraded due to the nonlinearity of the cos and min operation.
- The intensity image is the average of the RGB images.



a b c

FIGURE 6.49 HSI components of the noisy color image in Fig. 6.48(d). (a) Hue. (b) Saturation. (c) Intensity.